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EMPIRICALLY ADJUSTED WEIGHTED ORDERED P-VALUES METHOD WIMARSHA T. JAYANETTI, SINJINI SIKDAR AND N. RAO CHAGANTY DEPARTMENT OF MATHEMATICS AND STATISTICS OLD DOMINION UNIVERSITY

INTRODUCTION

- Meta analysis is used to integrate summary results from multiple studies targeting the same questions.
- One common practice is to use Fisher's [1] or Stouffer's [2] method for combining the p-values from multiple studies.
- The traditional methods aim at testing the alternative hypothesis that at least one of the studies is non-null.
- In recent years, researchers are more interested in identifying genes which are differentially expressed in majority of studies.
- WOP method [3] combines ordered p-values, weighting them based on their order, assuming p-values from individual studies are uniformly distributed under the null.
- In large-scale multiple testing, empirical distribution of p-values may not be uniform - so adjustments are needed.

PROPOSED METHOD

- Consider K independent studies where each study consisting of *G* genes.
- Let θ_{ij} denotes the underlying true effect size for the i^{th} gene in the j^{th} study, i = 1, 2..., G: j = $1, 2, \ldots, K.$

Hypothesis setting

For the i^{th} gene,

$$HS_m: \{H_0: \sum_{j=1}^K I(\theta_{ij} \neq 0) = 0 \, vs \, H_1^m: \sum_{j=1}^K I(\theta_{ij} \neq 0) \ge m\}$$

where $m = \lfloor K/2 \rfloor$, i.e., m is the smallest integer that is not lower than K/2.

Algorithm

- **Step 1**: For gene *i* in study *j*, obtain the p-value p_{ij} for testing the hypothesis of interest.
- **Step 2**: Consider the inverse z-transformation to get the corresponding z-scores as $\Phi^{-1}(p_{ij})$.
- **Step 3**: Let δ_0 and $\hat{\sigma}_0$ be estimated mean and standard deviation of the null distribution using central matching method [4]. Modify the z-scores, obtained in step 2 as:

$$z_{ij}' = \frac{z_{ij} - \hat{\delta}_0}{\hat{\sigma}_0}$$

SIMULATION STUDY

- Considered 10 studies each with 3000 genes.
- 50 genes considered to be differentially expressed (DE) in 1, 2, \cdots , 10 studies respectively.
- Considered 10% of the genes to be DE between two groups in at least 5 studies.
- Generated (log) gene expression values for the *i*th gene of the k^{th} subject in the j^{th} group for each experiment as:

$$y_{ijk} = \mu + G_i + V_j + GV_{ij} + W_{ijk} + e_{ijk}$$

where μ : overall mean effect, G_i : effect of the gene, V_j : effect of the group, GV_{ij} : interaction effect between gene and group, W_{ijk} : hidden confounder effect, e_{ijk} : random error term

- Set μ , G_i and V_j as zero for simplicity.
- Set the differences in magnitudes of DE genes between the two groups as 8 through the interaction term GV_{ij} .
- **Step 4**: Convert the empirically adjusted z-scores into corresponding p-values as:

$$p'_{ij} = \Phi(z'_{ij})$$

• **Step 5**: For a gene *i* order the p-values over the *K* independent studies. Let $p'_{i(j)}$ denote the j^{th} ordered p-value for gene *i*. Calculate the summary statistic as in [3] :

$$T_{i} = \sum_{j=1}^{K} w_{j} H(p_{i(j)}^{'})$$

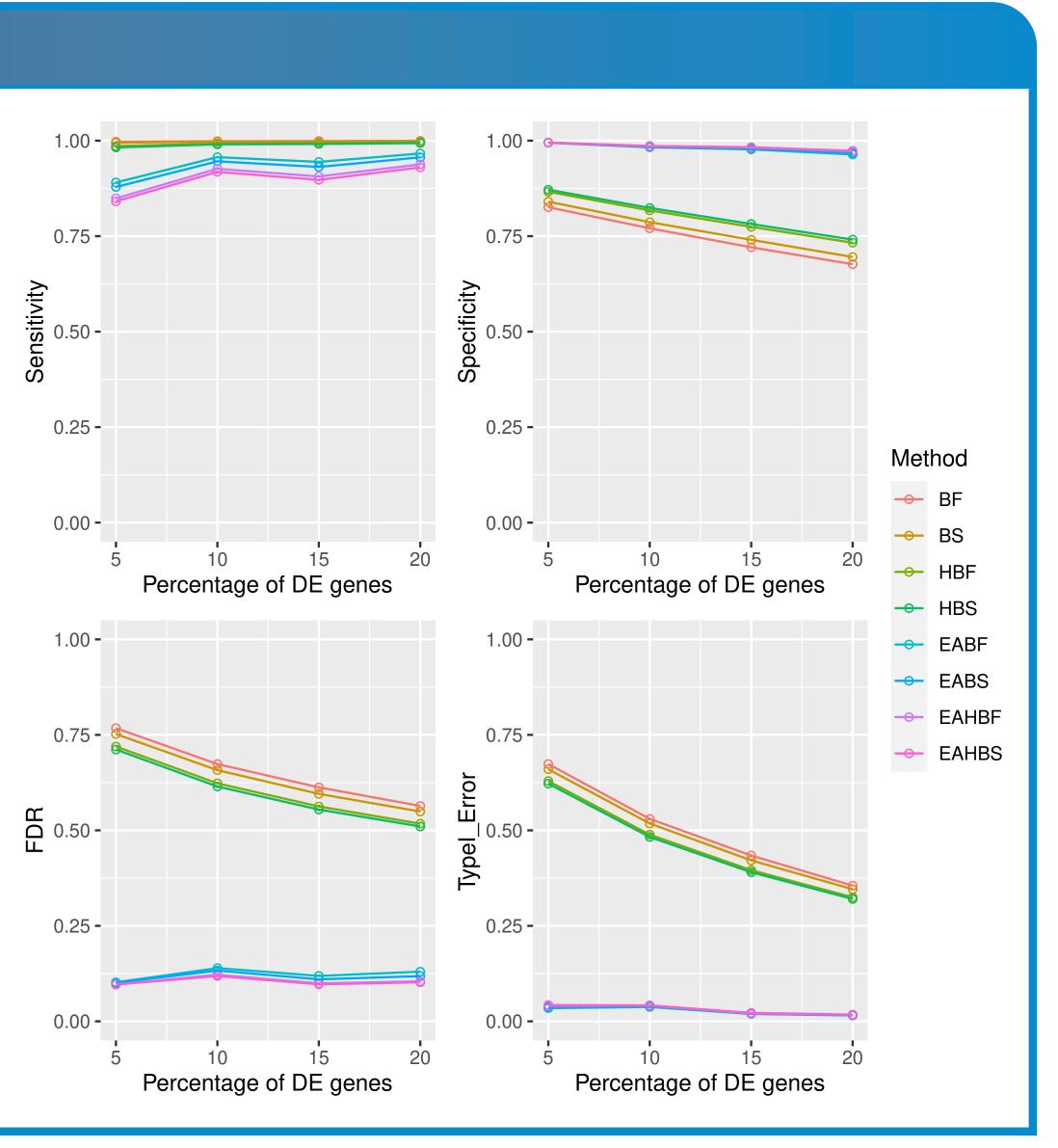
- Fisher's: $H(p_{i(j)}) = -2log(p_{i(j)})$
- Stouffer's: $H(p_{i(j)}) = \Phi^{-1}(1 p_{i(j)})$
- Binomial: $w_j^b = f(j 1: K 1, 0.5), j =$ 1, 2, ..., K, where f(x: n, p) denotes the pmf of Bin(n, p) for x = 0, 1, ..., n
- Half-binomial: $w_j^{hb} = w_j^b for m \leq j \leq$ K and 0 for j < m
- **Step 6**: For gene *i*, obtain p-value p^i by comparing the statistic, defined in step 5, to the numerical distribution by simulating U(0,1) random variables, i = 1, 2..., G.
- **Step 7**: Finally, apply the Benjamini-Hochberg method to account for multiple testing.

Method	Sensitivity	Specificity	FDR	Type I Error
BF	0.998	0.766	0.677	0.530
EABF	0.954	0.982	0.138	0.037
BS	0.997	0.783	0.660	0.517
EABS	0.944	0.984	0.129	0.039
HBF	0.992	0.814	0.626	0.488
EAHBF	0.925	0.985	0.119	0.041
HBS	0.989	0.821	0.618	0.482
EAHBS	0.916	0.986	0.116	0.041



• Generated hidden confounder as W_{ijk} = $u_{ijk}I(s_{ijk} = 1)$, where $s_{ijk} \sim Bernoulli(0.4)$ and u_{ijk} 's from Normal distribution such that effect of hidden confounder depends on gene ID, experiment ID and subject group.

• Compared performance of our Empirically Adjusted (EA) methods with original WOP methods. All results were averaged over 500 replicates.



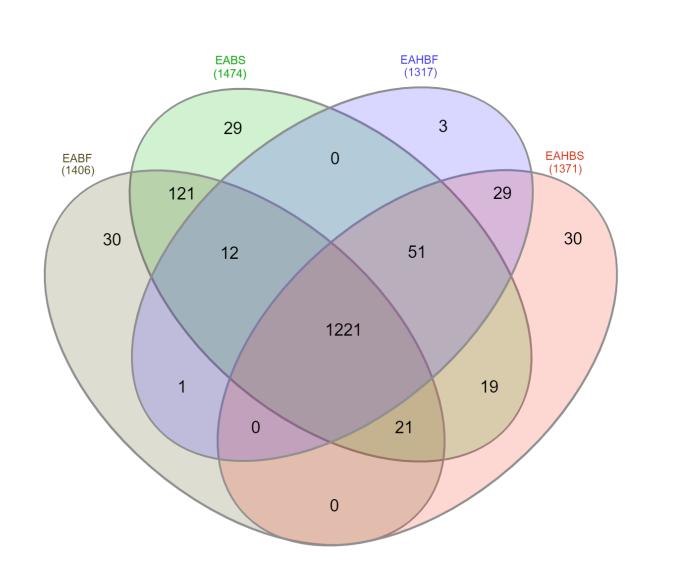
DATA ANALYSIS

• Identified DE genes between two lung cancer types.

• Data includes 5 studies with 7200 genes each.

 Tested the alternative hypothesis that genes are DE in at least 3 experiments out of 5 experiments (HS_3) .

Method	Number of DE genes (Percentage)			
wieniou	WOP	EAWOP		
BF	4921 (68.3%)	1406 (19.5%)		
BS	4672 (64.9%)	1474 (20.4%)		
HBF	4286 (59.5%)	1317 (18.3%)		
HBS	4208 (58.4%)	1371 (19.0%)		



• Pathway analysis identified biologically relevant pathways such as cell cycle, p53 signaling pathway, etc.

CONCLUSIONS

REFERENCES





• The proposed method has significantly better performance than the original WOP method especially in presence of hidden confounder.

• Type I errors are controlled at 5% for our methods while they are extremely high for the original WOP methods. FDR values are also significantly lower for the proposed methods.

• Our methods have slightly lower sensitivity values but much higher specificity values compared to the original methods.

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[2] Stouffer S.A., Suchman E.A., DeVinney L.C., Star S.A., and Williams Jr R.M. The american soldier: Adjustment during army life.(studies in social psychology in world war ii), vol. 1. 1949.

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[4] Efron B. Large-scale simultaneous hypothesis testing: the choice of a null hypothesis. *Journal of the American* Statistical Association, 99(465):96–104, 2004.